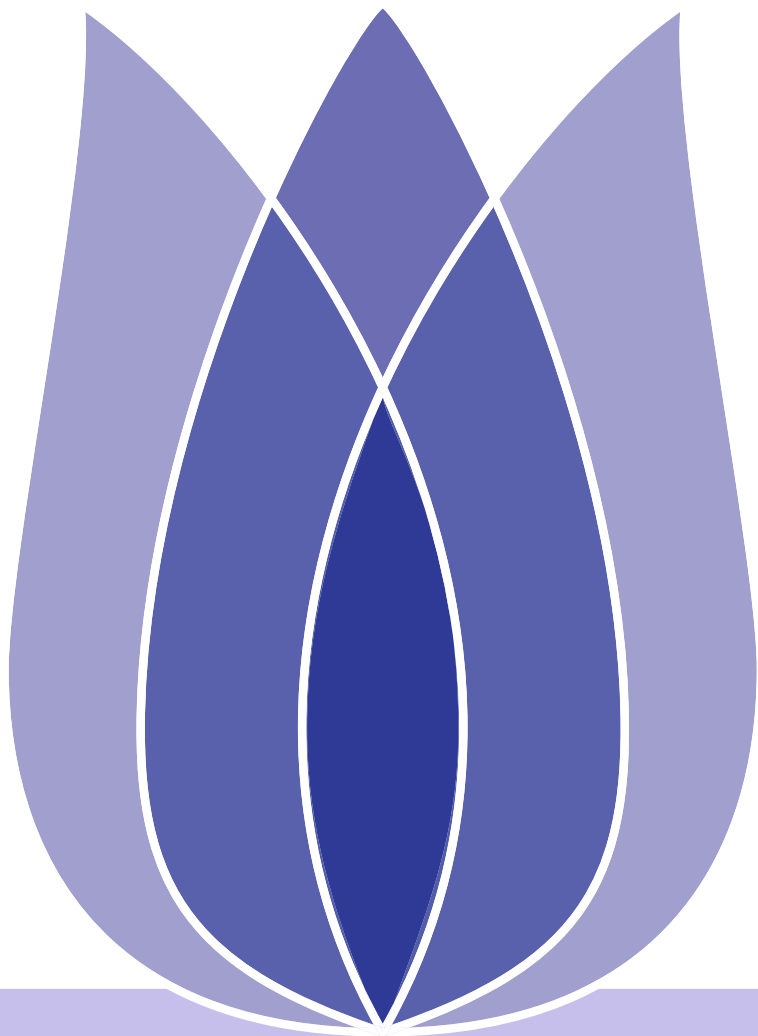


What's cooking?

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Problem

This time The Kaggle topic is What’s cooking?
The subject requirements are asks to predict the category of a dish’s cuisine given a list of its ingredients. There are three data sets:train.json,test.json,sample_submission.csv

Table 1: Data description

Name	Description	Attribute
train.json	training set(the type of cuisine, and the list of ingredients of each recipe)	Data: id, cuisine, ingredients
test.json	Test set(predict the cuisine type of the list ingredients)	Data:id,ingredients
sample_submission.csv	a sample submission file in the format	Data:id,cuisine



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Table 2: Data

	id	cuisine	ingredients
0	10259	greek	'romaine lettuce', 'black olives', 'grape tom...
1	25693	southern_us	'plain flour', 'ground pepper', 'salt', 'toma...
2	20130	filipino	'eggs', 'pepper', 'salt', 'mayonaise', 'cooki...
3	22213	indian	'water', 'vegetable oil', 'wheat', 'salt'
4	13162	indian	'black pepper', 'shallots', 'cornflour', 'cay...

- There are 39774 data in the training set.
There are 9944 data in the test set.
- Data is imported as DataFrame object, each recipe is a separate line.
- There are no missing values in the training set.



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- The percentage of dishes from each country in the total training set:

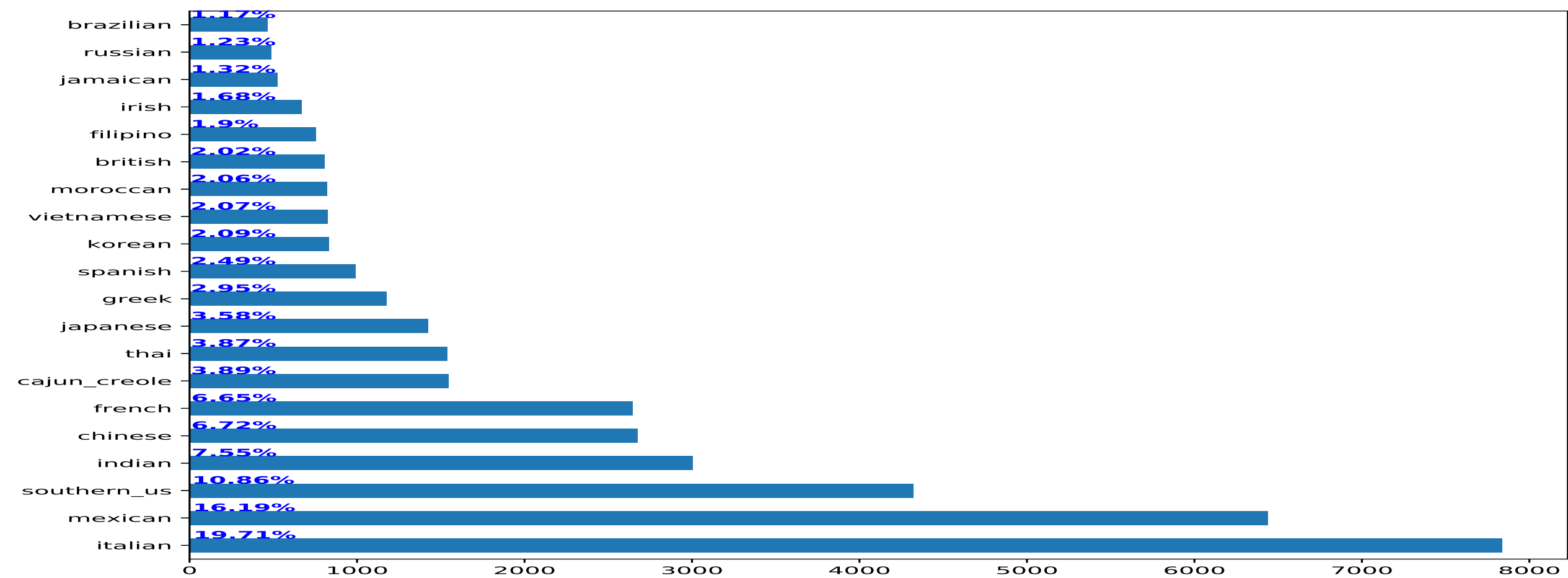


image1:The percentage of dishes



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- The following results are obtained by sorting and sorting the data of ingredients:

Table 3: Data

id	cuisine	id	cuisine	id	cuisine
salt	18049	sugar	6434	pepper	4438
onions	797	garlic cloves	6237	vegetable oil	4385
olive oil	7972	butter	4848	eggs	3388
water	7457	ground black pepper	4785	soy sauce	3296
garlic	7380	all-purpose flour	4632	kosher salt	3113

- We can also get the dish label from the training set,as follows:
'greek', 'southern_us', 'filipino', 'indian', 'jamaican', 'spanish', 'italian', 'mexican',
'chinese', 'british', 'thai', 'vietnamese', 'cajun_creole', 'brazilian', 'french', 'japanese',
'irish', 'korean', 'moroccan', 'russian'



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First process the string:

- To remove everything except a-z and A-Z and to make list element a string element.

Table 4: Data

id	cuisine	ingredients
0	greek	romaine lettuce black olive grape tomato garli...
1	southern_us	plain flour pepper salt tomato black pepper ...
2	filipino	egg pepper salt mayonaise cooking oil green ch...
3	indian	water vegetable oil wheat salt
4	indian	black pepper shallot cornflour cayenne pepper ...



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TFiDF Vectorizer on the processed data:

- Use TFiDF Vectorizer to evaluate the importance of each dish of vegetable raw materials

(0, 605)	0. 006881830426943897
(0, 1598)	0. 004922916349018388
(0, 1217)	0. 009471373982023997
(0, 515)	0. 004407654548451798
(0, 697)	0. 0059309209813973975
(0, 2003)	0. 008325482704302814
(0, 1947)	0. 008325482704302814
(0, 783)	0. 00751245990409849
(0, 1727)	0. 004922916349018388
(0, 532)	0. 006366568626377306
(0, 734)	0. 005220677348656123
(0, 1670)	0. 009471373982023997
(0, 1673)	0. 016660637478518942
(0, 2134)	0. 013763660853887794
(0, 945)	0. 013763660853887794
(0, 110)	0. 004922916349018388
(0, 700)	0. 006366568626377306
(0, 357)	0. 006366568626377306

image4:TFiDF



similar dishes

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More similar dishes and visualize dishes:
From this picture, We can notice there are 3 clusters of cuisines.

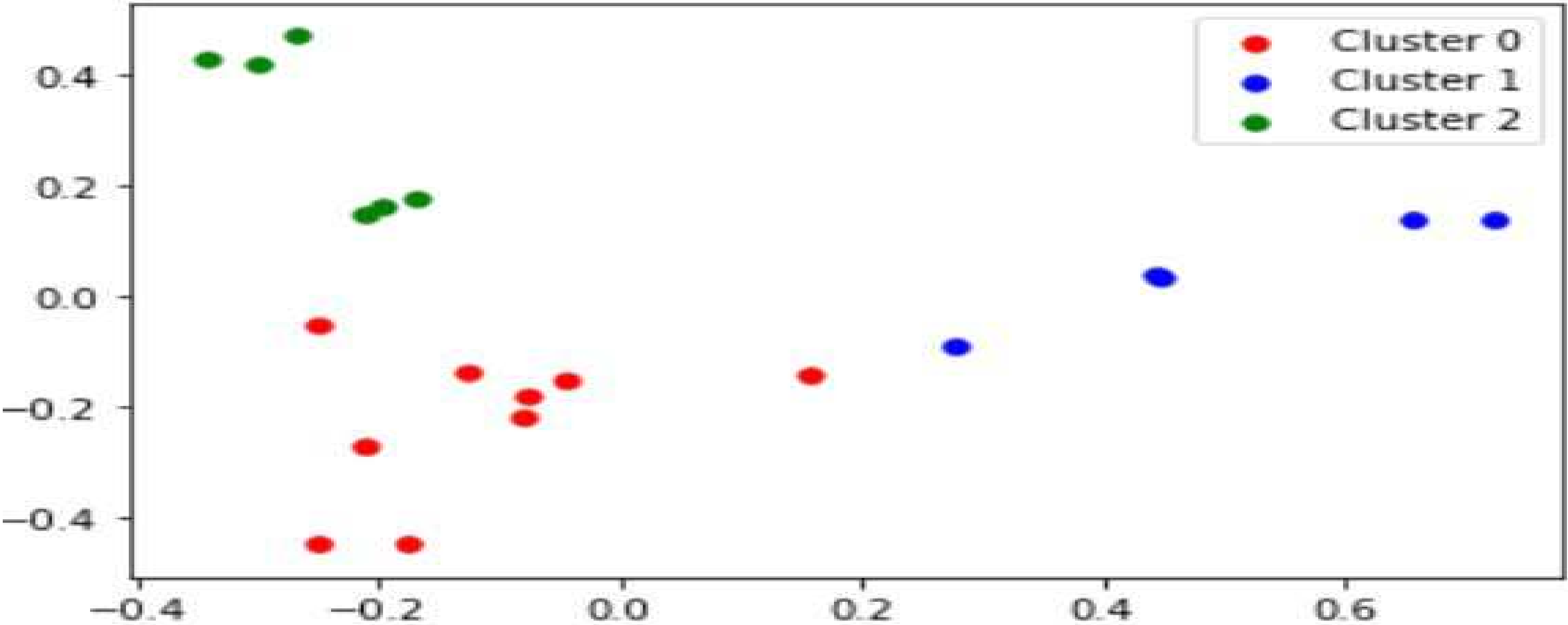


image2:More similar dishes



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Model

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Use Linear SVC model for prediction:

■ best_score:0.7857898061559815